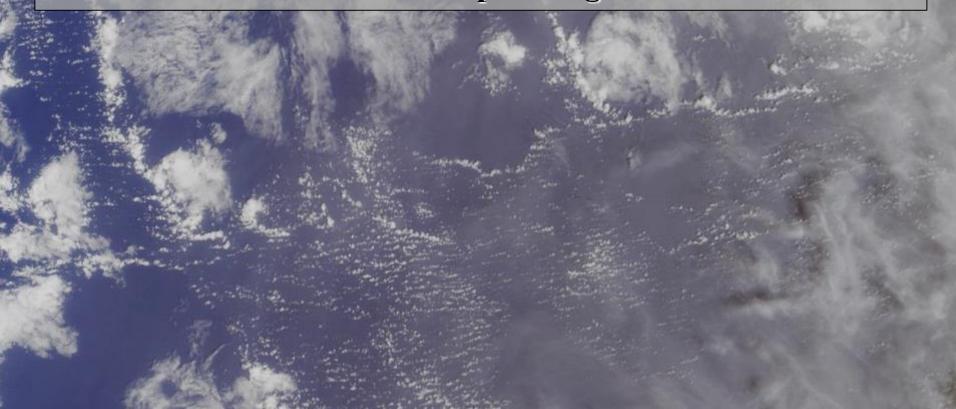
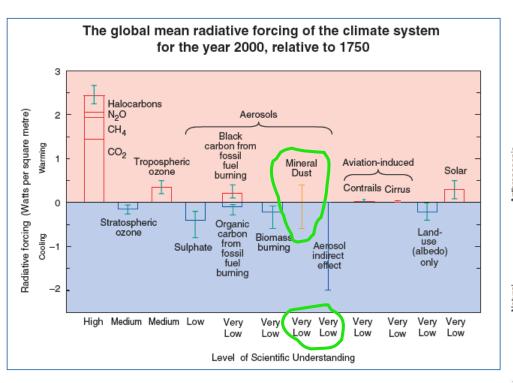
Remote Sensing of Aerosols From Satellites:
Why has it been do difficult to quantify
Aerosol-Cloud Interactions for climate assessment,
and How can we make Progress?

Ralph Kahn NASA/Goddard Space Flight Center



Even DARF and Anthropogenic DARF are *NOT* Solved Problems (Yet)



Radiative Forcing Components

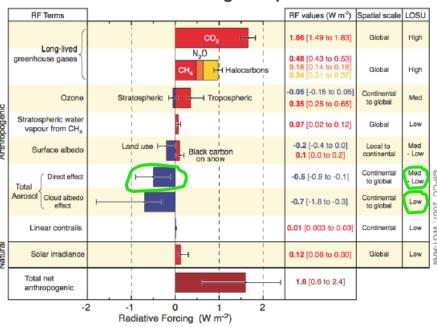
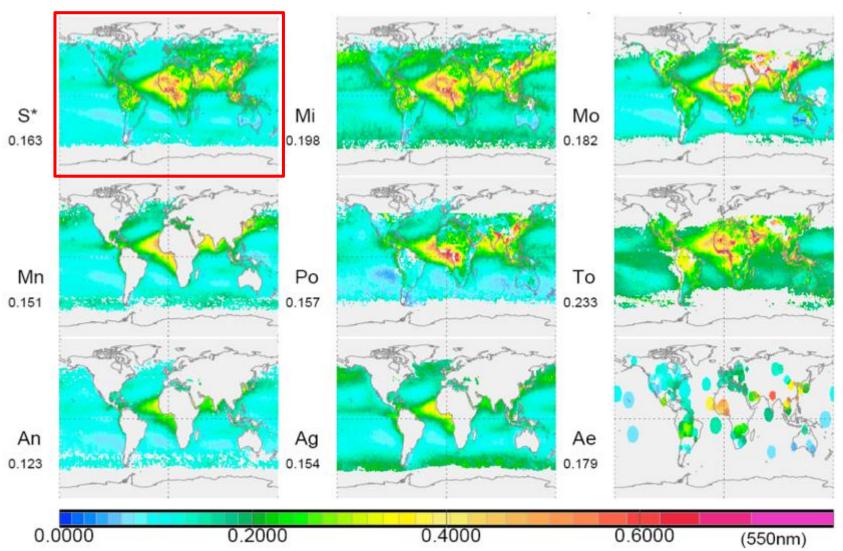


FIGURE SPM-2. Global-average radiative forcing (RF) estimates and ranges in 2005 for anthropogenic carbon dioxide (CO₂), methane (CH₄), nitrous oxide (N₂O) and other important agents and mechanisms, together with the typical geographical extent (spatial scale) of the forcing and the assessed level of scientific understanding (LOSU). The net anthropogenic radiative forcing and its range are also shown. These require summing asymmetric uncertainty estimates from the component terms, and cannot be obtained by simple addition. Additional forcing factors not included here are considered to have a very low LOSU. Volcanic aerosols contribute an additional natural forcing but are not included in this figure due to their episodic nature. Range for linear contrails does not include other possible effects of aviation on cloudiness. {2.9, Figure 2.20}

IPCC AR3, 2001 (Pre-EOS)

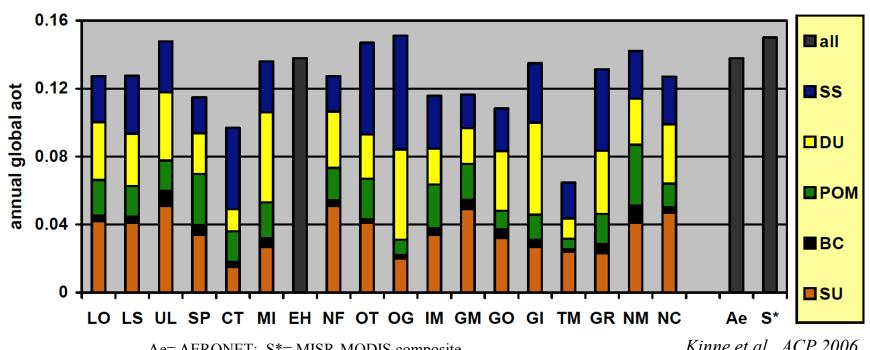
IPCC AR4, 2007 (EOS + ~ 6 years)

Multi-year Annual Average *Aerosol Optical Depth* from Different Measurements + *Synthesis* (S*)



From: Kinne et al. ACP 2006

Constraining ARF – The Next Big Challenge



Ae= AERONET; S*= MISR-MODIS composite

Kinne et al., ACP 2006

Note: These are <u>not</u> yet updated to the CMIP5 (AR5) models

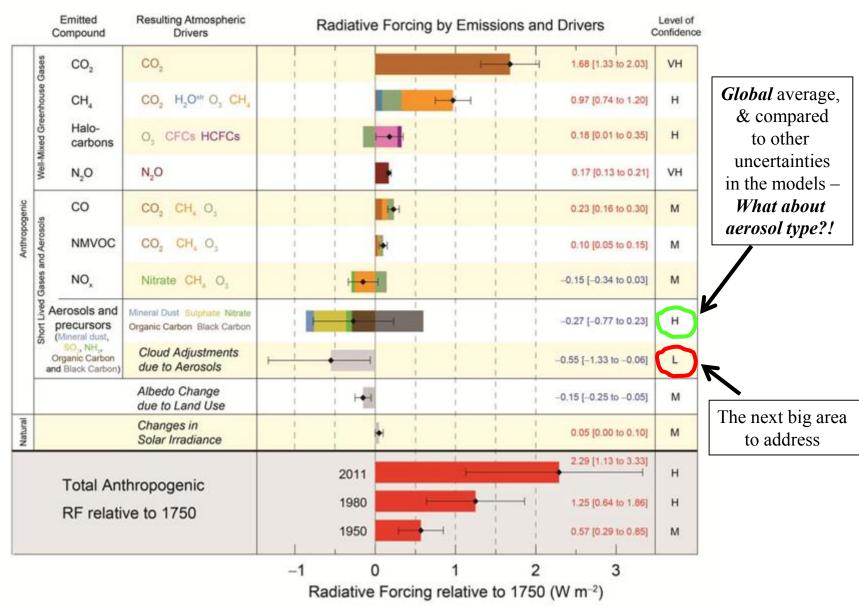
• The next big observational challenge: Producing *monthly*, *global maps of* <u>Aerosol Type</u>

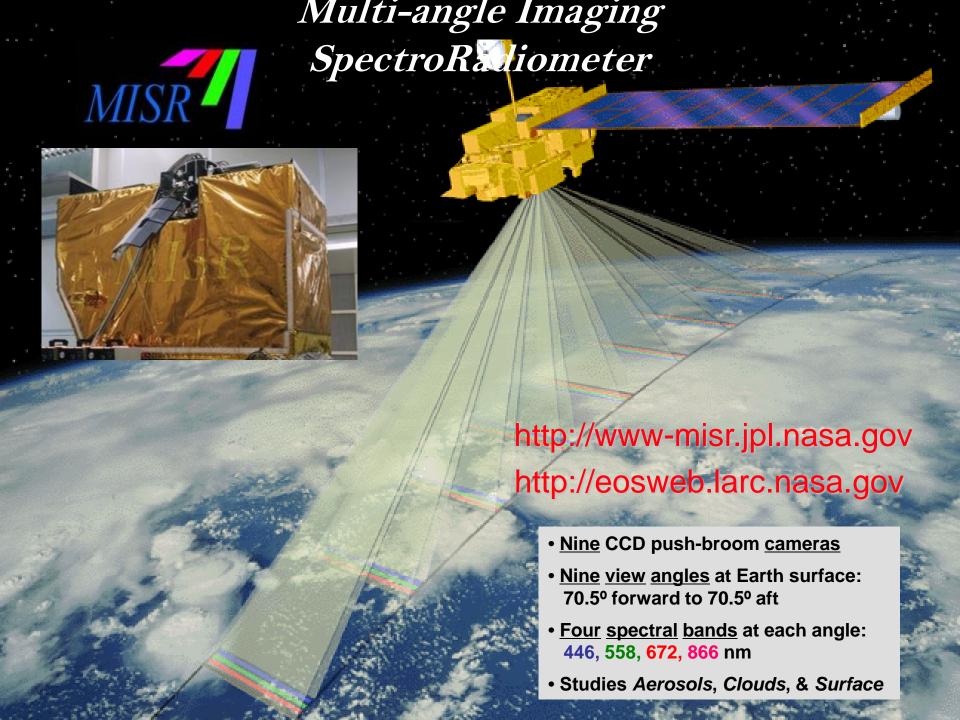
How Good is *Good Enough*?

Instantaneous AOD & **SSA** uncertainty upper bounds for $\sim 1 \text{ W/m}^2 \text{ TOA DARF}$ accuracy: ~ 0.02

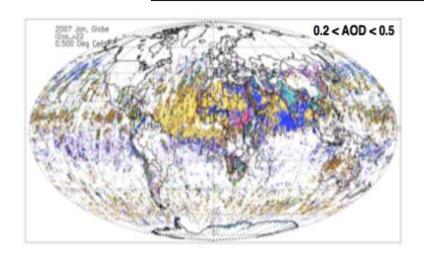
-- For aerosol indirect effects, the aerosol type constraint requirements are more stringent

The Current Assessment of Climate Forcing Factors

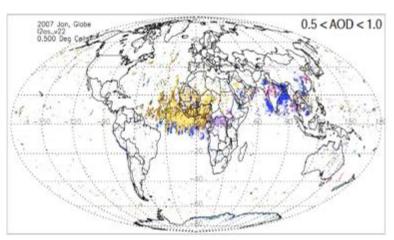


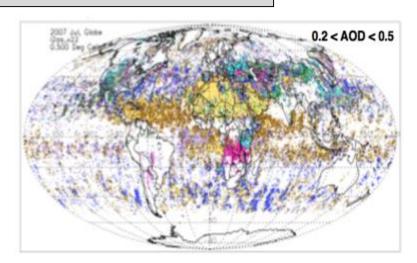


MISR Aerosol Type Discrimination

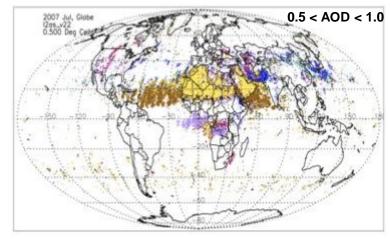


January 2007





July 2007



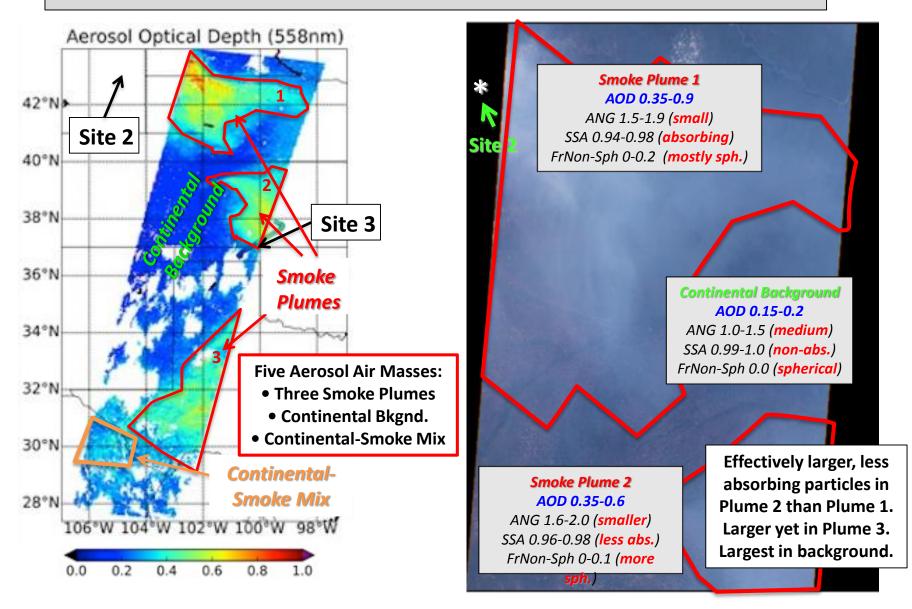
Mixture Group

1-10 11-20 21- 31-40 41-50 51-62 63-70 71-74

Spherical, non-absorbing

Non-spherical

SEAC⁴RS – MISR Overview 19 August 2013



Passive-remote-sensing *Aerosol Type* is a *Total-Column-Effective*, *Categorical* variable!!

For Aerosol-Cloud Interactions — Overall Satellite *Limitations*

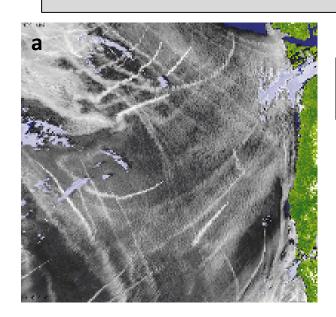
- Polar orbiters provide snapshots only
- Difficult to probe *cloud base*
- Typically ~100s of meters or poorer *horizontal resolution*
- Passive instruments (imagers) offer little vertical information
- Active instruments (e.g., lidar) offer little spatial coverage
- Little information about aerosol *particle microphysical properties*
- Bigger issues retrieving aerosols in the presence of clouds!
- Cloud property retrievals can be aliased by the presence of aerosols

Finer Points on Satellite Aerosol Retrieval *Limitations*

- Difficult to retrieve aerosols that are *collocated with cloud* -- Cloud-scattered light & cloud "contamination" can affect near-cloud aerosol retrievals
- Rarely can detect aerosol in *droplet-formation region* below clouds – need cloud & aerosol vertical distributions
- Aerosols smaller than about **0.1 micron diameter** look like atmospheric gas molecules – must *infer CCN* number
- Must deduce aerosol *hygroscopicity* (composition) from qualitative "type" – size, shape, and SSA constraints
- Environmental (Meteorological) Coupling Factors can *co-vary*
 - -- LWP can decrease as aerosol number concentration increases (also depends on atm. stability)
- Many aerosol-cloud interaction time & spatial scales do not match *satellite sampling*

Satellites are fairly blunt instruments for studying aerosol-cloud interactions!!

Aerosol Effects on Clouds – 'Controlled' Situations



FREQUENCY 0.30 0.20

RELATIVE

FREQUENCY

0.10

0.20

0.15

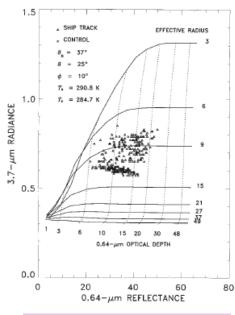
0.10

0.05

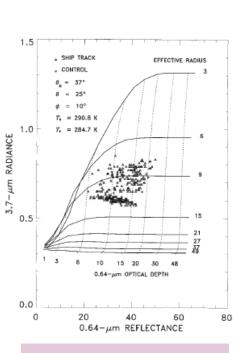
 Δr_{c}

 $\Delta \tau_{c}$

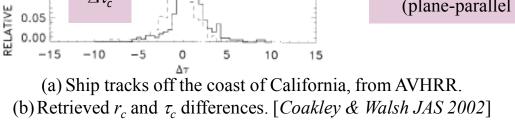
- Statically stable conditions
- Fairly uniform stratiform



 τ_c and r_c from 0.64 and 3.7-micron AVHRR (plane-parallel RT)



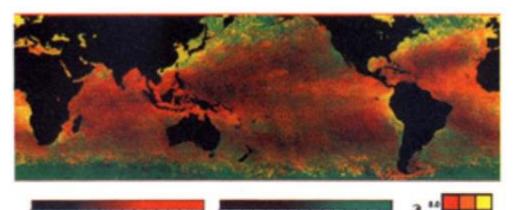
False-color AVHRR [Blue – 11 μm; $Red - 0.67 \mu m; Green - 3.7 \mu m$ Red indicates large droplets, yellow signifies smaller droplets [Rosenfeld, Sci. 2000]



5

 $\Delta r_* (\mu m)$

Aerosol Effects on Clouds – Correlation Studies

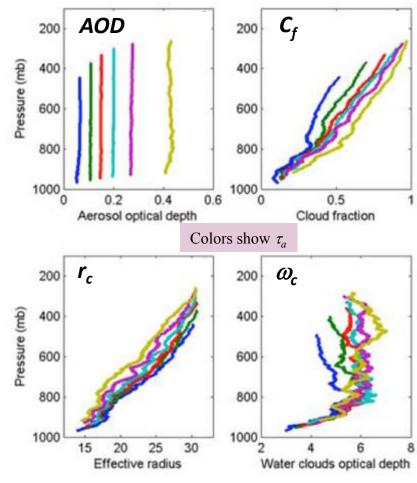


$\operatorname{Log} N_a$ (red) vs. $\operatorname{Log} N_c$ (green)

Yellow = N_a , N_c large Red = N_a large, N_c small Green = N_a small, N_c large

Correlation between AVHRR particle number N_a (fixed r_a ; AI (= τ_a x ANG) and cloud droplet (N_c) concentrations, for 4 months in 1990; $N_a \sim \tau_c$; $N_a \sim 1/r_c$ in low cloud (yellow) regions [Nakajima et al., GRL 2001].

- [Feingold et al. JGR 2001] **Drop size effect** saturates at $\tau_a \sim 0.4$, 0.8, depending on conditions (SCAR-B, Brazil)
 - [Ackerman et al., Sci. 2000] INDOEX absorbing aerosol can dissipate clouds

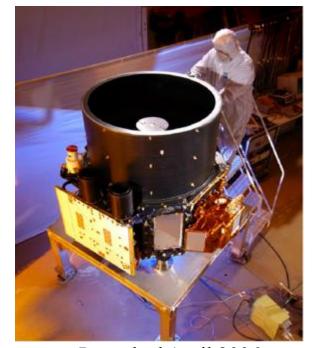


Atlantic convective *cloud invigoration* from MODIS [Koren et al. GRL 2005]

- $1/r_c \sim N_c \sim N_a \sim \tau_a$ [Cloud radius effect]
- r_c decrease \rightarrow early precip. inhibited \rightarrow higher cloud tops, cloud fraction, glaciation
- C_f , T_c , τ_c (water clouds) all increase with τ_a

The Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observations (CALIPSO)

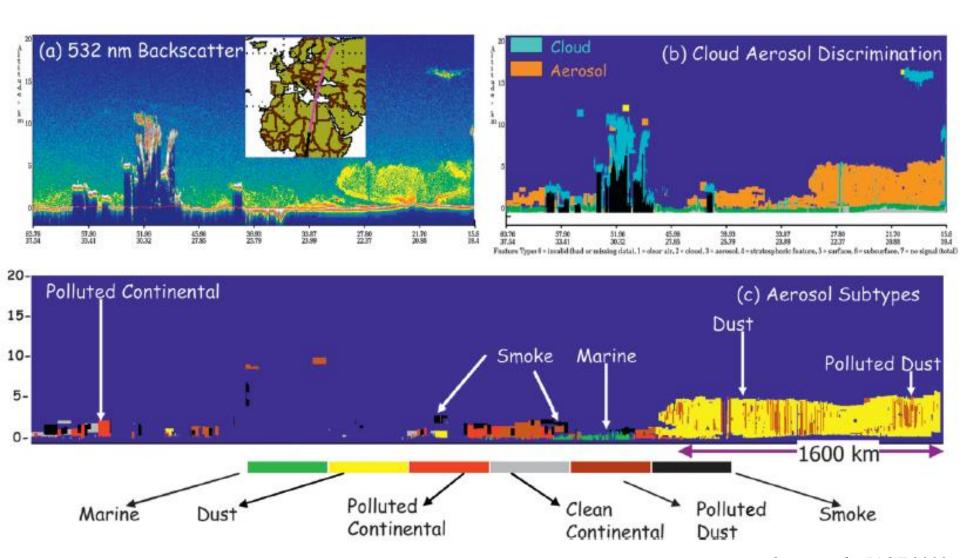
Vertical Range (km)	Horizontal Resolution (km)	Vertical Resolution (m)	
30.1 – 40	5	300	
20.2 - 30.1	1.7	180	
8.2 – 20.2	1.	60	
-0.5 - 8.2	0.33	30	

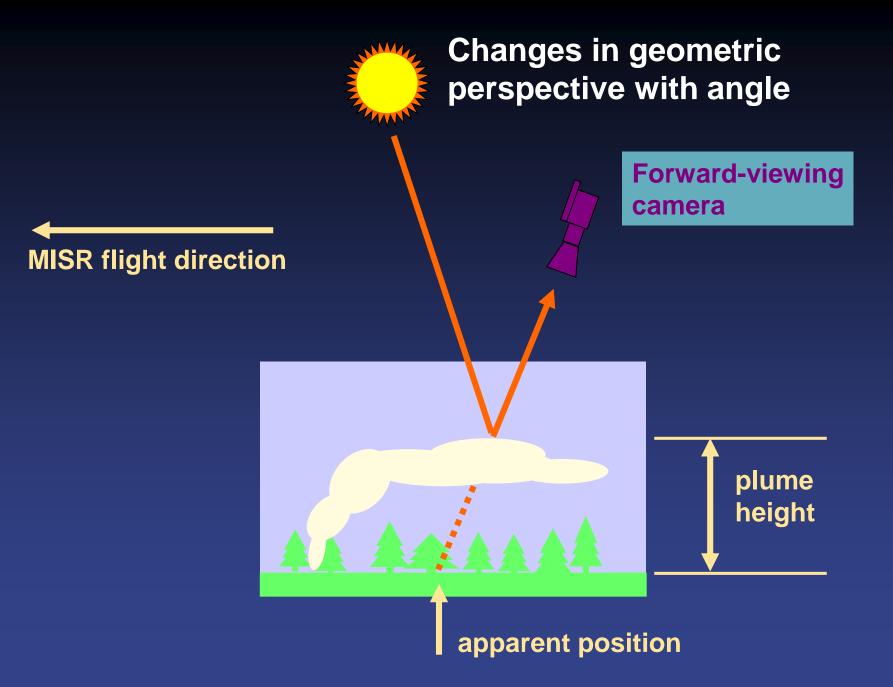


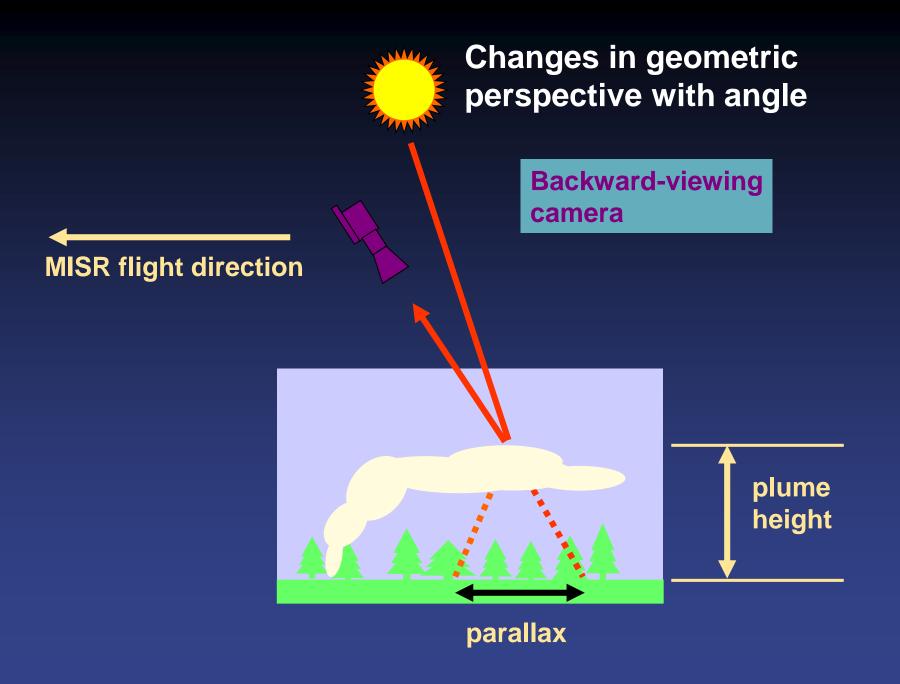
Launched April 2006

- 15 orbits per day, ~100 m wide sampling *curtain*; averaged to 333 m
- 532 and 1064 nm + polarization (at 532 nm); to ~40 km elevation
- Layer height for AOD $\geq 10^{-2}$; AOD for layers having AOD ≤ 3
- For low AOD, need the higher S/N of *nighttime*, 532 nm observations

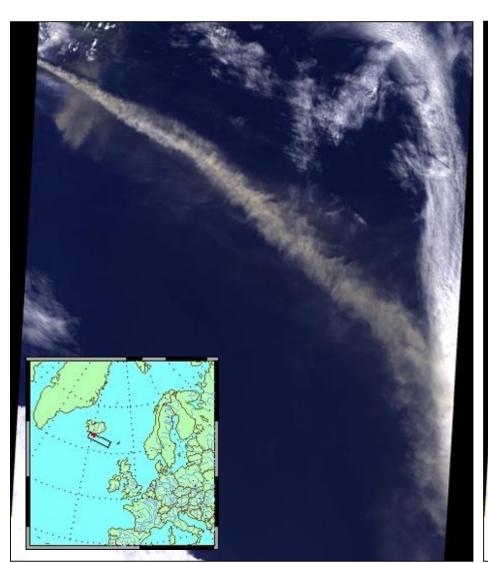
The Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observations (CALIPSO)

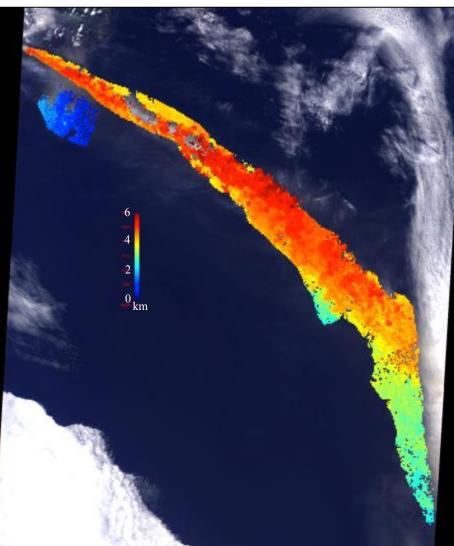






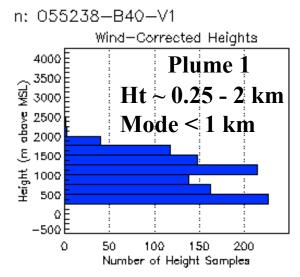
MISR Stereo-Derived Plume Heights 07 May 2010 Orbit 55238 Path 216 Blk 40 UT 12:39

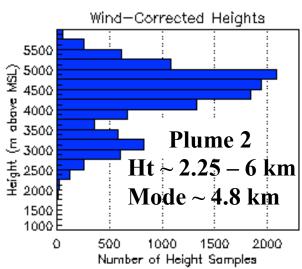




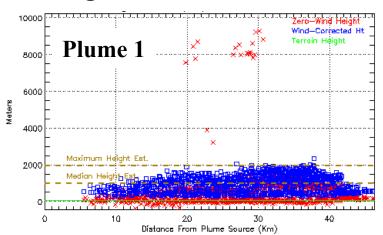
D. Nelson and the MISR Team, JPL and GSFC

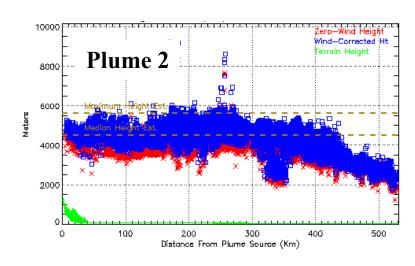
MISR Stereo-Derived Plume Heights 07 May 2010 Orbit 55238 Path 216 Blk 40 UT 12:39





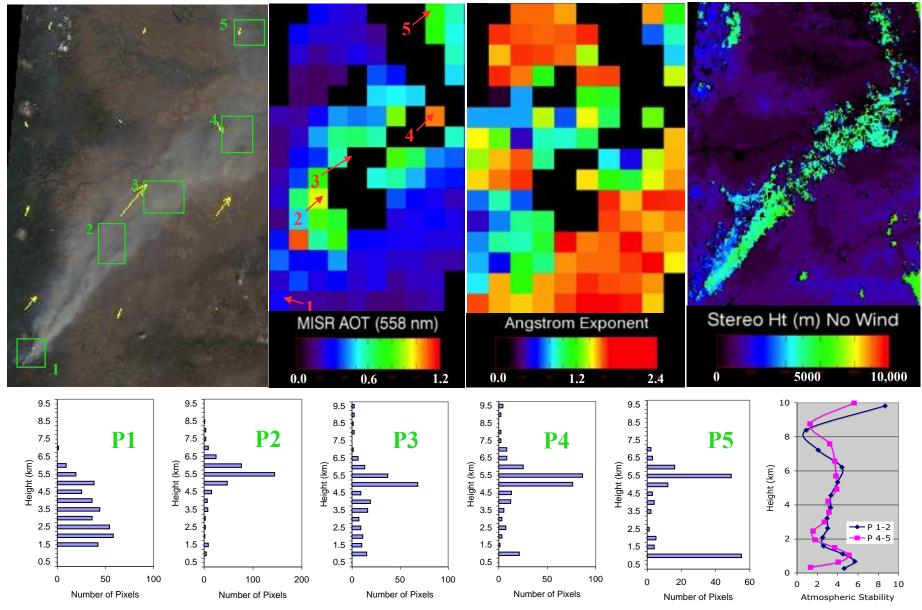
Height: Blue = Wind-corrected





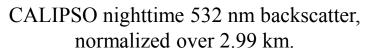
Oregon Fire Sept 04 2003

Orbit 19753 Blks 53-55 MISR Aerosols V17, Heights V13 (no winds)

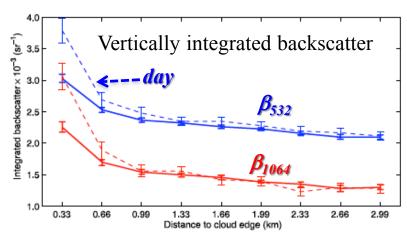


Kahn, et al., JGR 2007

Aerosol Properties Near Cloud

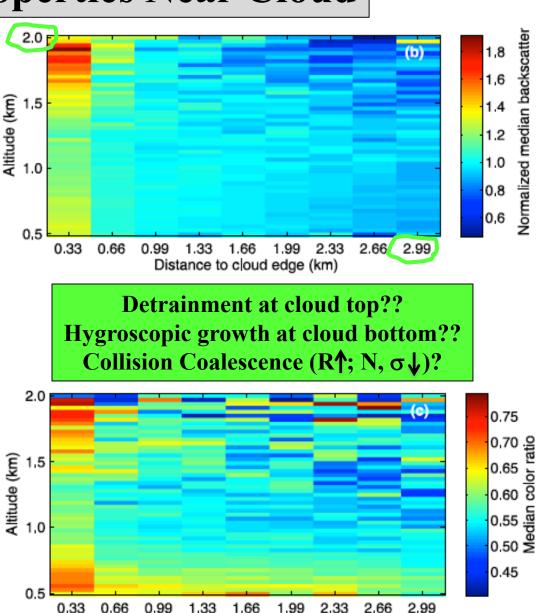


Enhanced aerosol opacity near cloud edge, especially at **cloud top and bottom**.



CALIPSO median nighttime 1064/532 nm color ratio. **Larger particles** near cloud edge,

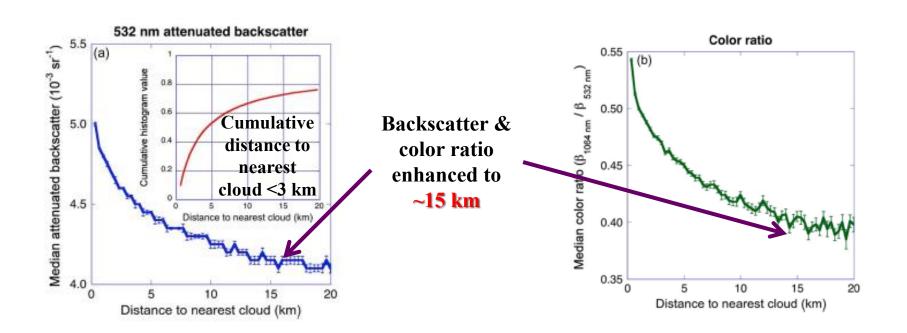
especially at cloud top and bottom.



Distance to cloud edge (km)

Tackett & Di Girolamo GRL 2009

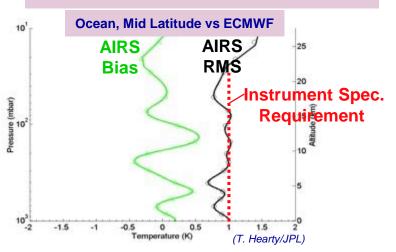
Aerosol Properties Near Cloud



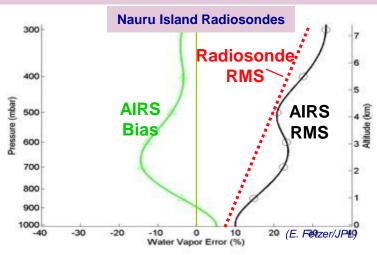
Global data Sept. – Oct. 2008

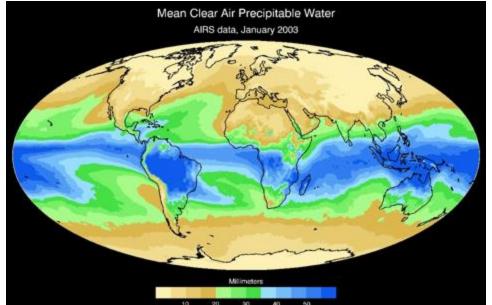
AIRS - Temperature & Water Vapor Profiles

Temperature Profiles Accurate to 1K/km to 30 mb



Water Vapor Profiles Match Observations 15%/2km





15 km nadir footprint

Satellite Capabilities

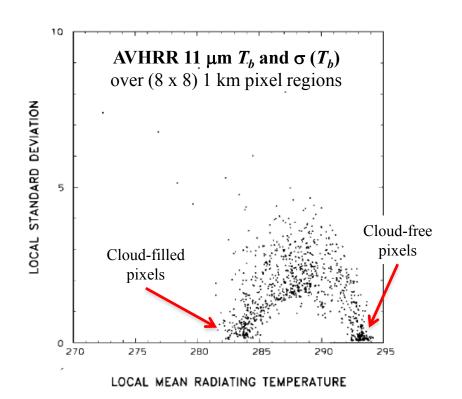
- Polar orbiting imagers provide frequent, global coverage
- Geostationary platforms offer high temporal resolution
- Multi-angle imagers offer aerosol plume height & cloud-top mapping
- Passive instruments can retrieve total-column aerosol amount (AOD)
- Active instruments determine aerosol & some cloud vertical structure
- UV imagers and active sensors can retrieve aerosol above cloud
- Multi-angle, spectral, polarized imagers obtain some aerosol type info.
- Active sensors can obtain some aerosol type info., day & night
- Satellite trace-gas retrievals offer *clues about aerosol type*
- Vis-IR imagers can retrieve cloud phase, r_c , T_c , p_c , τ_c , α_c , C_f , LWP

Need to be creative & Play to the strengths of what satellites offer!!

Assessing Some Satellite-Retrieval Issues

Partly Filled Pixels

[Coakley & Bretherton JGR 1982]



- Can obtain cloud-fraction for single-layer clouds
- Multi-layered clouds can be identified by distinct T_b
- The challenge is selecting a spatial scale for aggregation

Sampling Bias Example

[Rosenfeld & Feingold GRL 2003]

First Indirect Effect: $IE \sim -d \ln r_c / d \ln \tau_a$

AVHRR

[*IE* ~ **0.17**] **over ocean** (*Nakajima et al. 2001*)

- Partly filled pixels, surface contributions $\rightarrow r_c$ errors
- Disfavors: thin & broken cloud, especially over land

POLDER (Breon et al., 2002) [IE \sim 0.085] over ocean; [IE \sim 0.04] over land

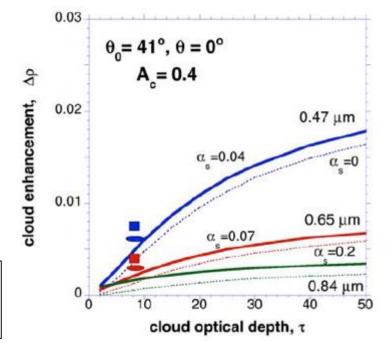
- Uses "glory" to get $r_c \rightarrow$ favors more mono-disperse, less turbulent clouds
- Disfavors: thick convective clouds, variable height & r_c

Thinner clouds →
smaller updrafts, less activation, smaller *IE*So POLDER might produce artificially low regional *IE*

3-D Light Scattering Effects on Remote Sensing



ASTER false-color image *Brazil, 09 August, 2001*



Refl. in "clear" pixels used for MODIS AOD Retrievals (*squares*)

Refl. in pixels 3 km away from cloud (*ovals*) [*Wen et al.* 2007]

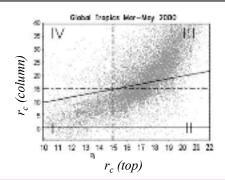
Simulated cloud \rightarrow Rayleigh scattered light enhancement vs. τ_c

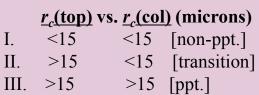
- Using the image geometry
 - For three wavelengths
- For different surf. reflectances (α_s)

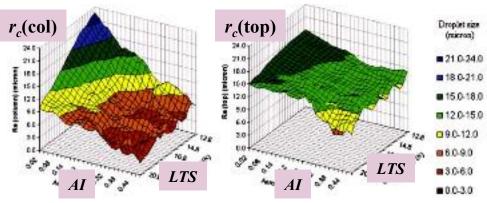
Vertical Structure, and Confounding Meteorology r_c – Cloud 'Top' vs. Cloud Column, & *LTS*

- TRMM data, March-May, 2000; 37°N to 37°S
- Vis-IR Radiance Imager (VIRS) for $r_c(top)$, τ_c
- Microwave Imager (TMI) for $r_c(col)$, LWP (19, 37GHz)
- Warm clouds only $(T_c > 273 \text{ K})$
- VIRS to find cloud-filled TMI pixels
- AI from MODIS
- Lower Trop. Stability (*LTS*) from NCEP
- *IE* appears larger for $r_c(col)$ than $r_c(top)$
- Higher LTS and/or $AI \sim$ reduced r_c and suppressed rain conditions
- Aerosol effect ~ 50% larger than *LTS* effect
- TMI *LWP* decreases with reduced $r_c \rightarrow$ net change in cloud albedo SMALL

[$d\alpha_c/dLTS \sim 9\%$; LTS effect dominates]

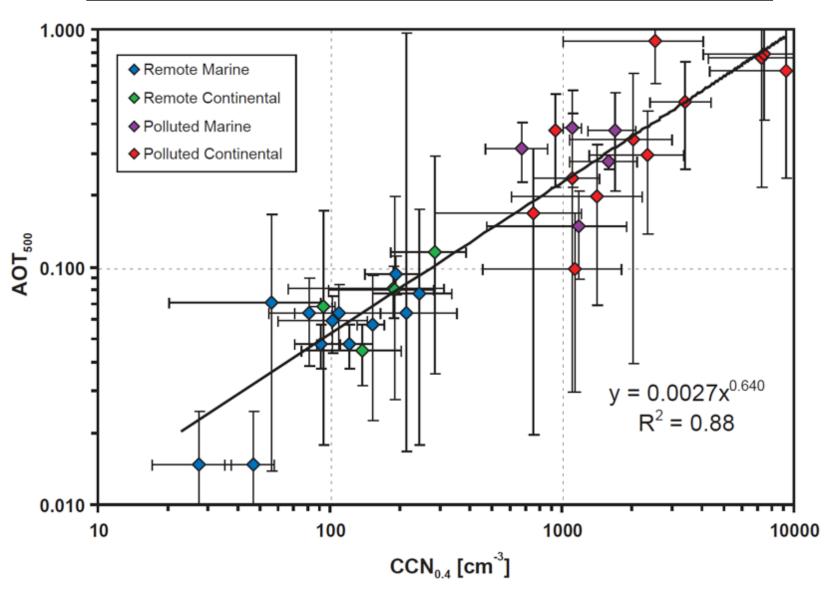






r_c vs. AI vs. LTS

Correlation Between AOD from Space and CCN in Remote & Polluted Regions



Using $AI (= \tau_a \times ANG)$ to Estimate CCN

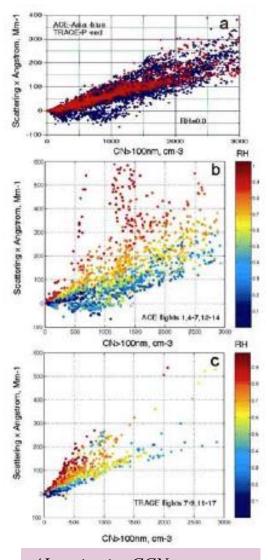
Kapustin, Clarke, et al., JGR 2006

- <u>Test Idea</u>: Smaller particles more likely to become *CCN*; *Ang* is a smaller quantity for larger particles
- ACE-Asia, Trace-P in situ field data CCN proxy
- AI does not work quantitatively in general, but can if the data are stratified by:
- -- **RH** in the aerosol layer(s) observed by satellites
- -- Aerosol Type (hygroscopicity; pollution, BB, dust)
- -- Aerosol Size (Ang is not unique for bi-modal dist.)

Practically, in addition to t_a and Ang, this requires:

- -- Vertical humidity structure
- -- Height-resolved aerosol type
- -- **Height-resolved size** dist. [extrapolated to small sizes(?)]

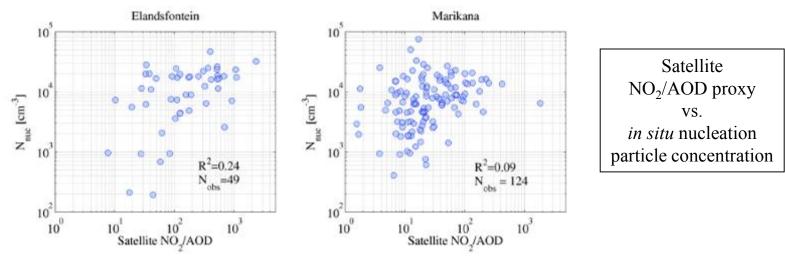
This study includes enough detail to assess $AI \sim N_a$ and $AI \sim CCN$



AI vs. in situ CCN proxy
(a) all ACE (blue) & Trace-P, dry
(b) ACE - OPC-only, amb. RH

(c) TP - OPC-only, amb. RH

Satellite-Derived Proxies for CCN



• OMI NO₂ Column

Sundstrom et al., ACP 2015

- OMI *SO*₂ Column (mainly near-surface)
- OMI UVB (310 nm) Surface noontime irradiance to form secondary sulfate
- MODIS AOD [attempt to represent the condensation sink for nucleation particles]

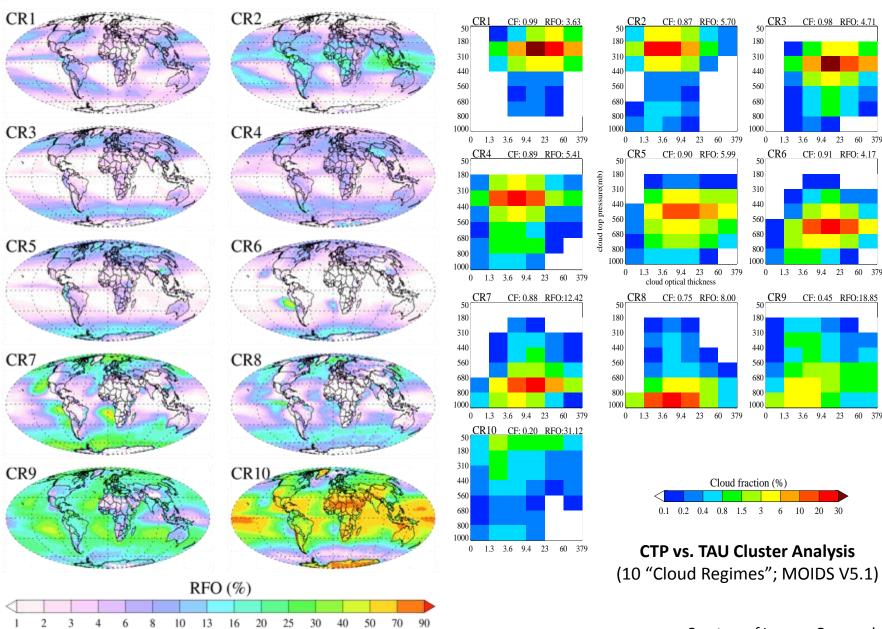
These are quantities we can retrieve from satellites, though they are not necessarily the ones we really want

- →Ambiguity in vertical distributions of formation areas and sinks
- → Lack of information about diurnal variation from satellites
- → The 2-D spatial distribution of proxies compares ~ better with *in situ* observations for S. Africa, except where gas column concentrations are low

Would you believe the answer if it were a surprise?



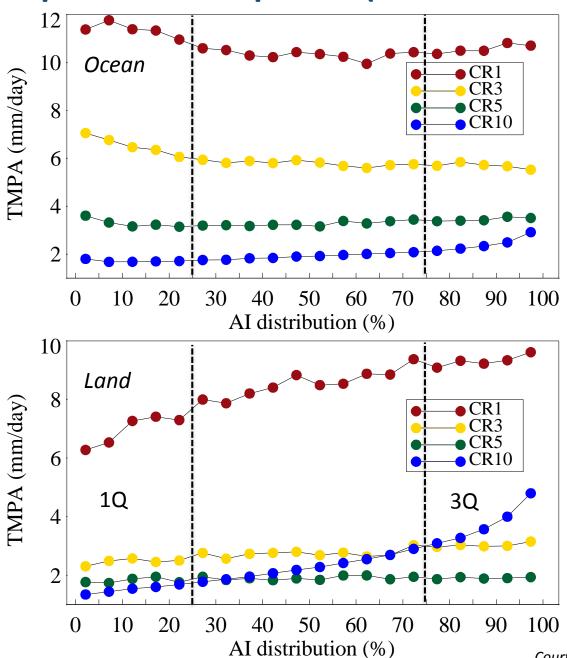
MODIS global cloud regimes



Frequency of Occurrence



Precipitation vs Al per CR (50° S to 50° N)



Relationship
between
precipitation &
Aerosol Index,
stratified by
cloud regime (CR)
and Land/Ocean

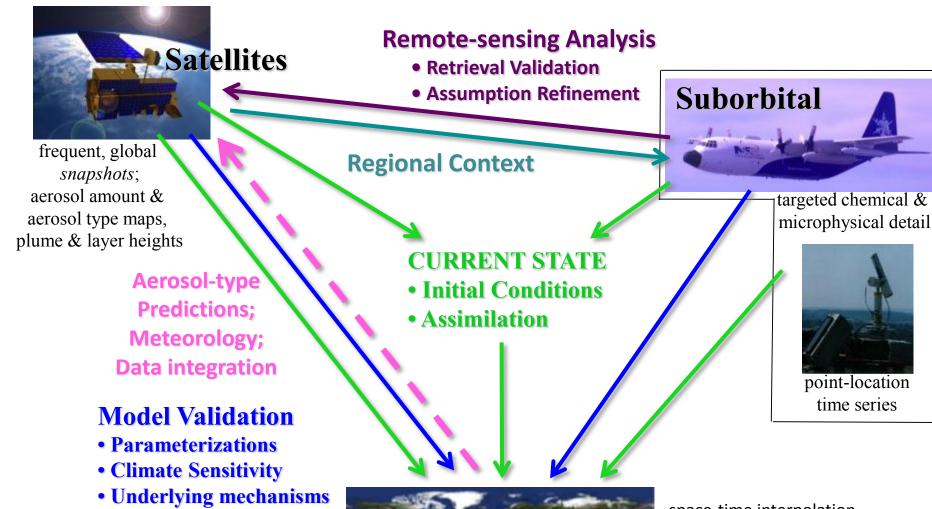


Summary

Observed trends when going from low aerosol index (1Q) to high (3Q)

	Land/	ice Ocean , 2, 3)	CR _{liq} Land/Ocean (CR 6, 7, 8)		CR ₁₀
Precip.	1	₩		-	1
C_f	-	-	1	<u> </u>	1
СТН	1	1	1	-	1
$ au_c$	1	U	1		1
r _e	U	-	1	ļ	1
PrecipNZ	1	U	-	1	1

red arrow: consistent with invigoration; blue arrow: consistent with 1st and 2nd indirect effect



Models

Must *stratify* the global satellite data to treat appropriately situations where different physical mechanisms apply

space-time interpolation, **Aerosol Direct & Indirect Effects** calculation and prediction

point-location time series

Adapted from: Kahn, Survy. Geophys.

SAM-CAAM

[Systematic Aircraft Measurements to Characterize Aerosol Air



[This is currently a *concept-development effort*, not yet a project]

Primary Objectives:

- Interpret and enhance 15+ years of satellite aerosol retrieval products
- Characterize statistically particle properties for major aerosol types globally,

to provide detail unobtainable from space, but needed to *improve*:

- -- Satellite aerosol retrieval algorithms
- -- The translation between satellite-retrieved aerosol optical properties

SAM-CAAM Concept

[Systematic Aircraft Measurements to Characterize Aerosol Air Masses]

- **Dedicated Operational Aircraft** routine flights, 2-3 x/week, on a continuing basis
- Sample Aerosol Air Masses accessible from a given base-of-operations, then move; project science team to determine schedule, possible field campaign participation
- Focus on in situ measurements required to characterize particle Optical Properties, Chemical Type, and Mass Extinction Efficiency (MEE)
- *Process Data Routinely* at central site; instrument PIs develop & deliver algorithms, upgrade as needed; data distributed via central web site
- Peer-reviewed Paper identifying *4 Payload Options*, of varying ambition; subsequent selections based on agency buy-in and available resources

SAM-CAAM is feasible because:

Unlike aerosol amount, *aerosol microphysical properties tend to be repeatable* from year to year, for a given source in a given season